

## EAST Search History

| Ref # | Hits | Search Query   | DBs   | Default Operator | Plurals | Time Stamp       |
|-------|------|--|---|------------------|---------|------------------|
| S1    | , 1  | "017504".apn.  | US-PGPUB;<br>USPAT;<br>USOCR;<br>EPO; JPO;<br>DERWENT;<br>IBM_TDB | OR               | OFF     | 2007/07/29 08:33 |
| S2    | 116  | ((cluster\$3 or group\$3) with categor\$3 with (structur\$3 or tree or hierarch\$3)).ab.             | US-PGPUB;<br>USPAT;<br>USOCR;<br>EPO; JPO;<br>DERWENT;<br>IBM_TDB | OR               | OFF     | 2007/07/29 09:53 |
| S3    | 354  | ( weight\$3 or scor\$3 or rank\$3 ) with vector with categor\$3                                      | US-PGPUB;<br>USPAT;<br>USOCR;<br>EPO; JPO;<br>DERWENT;<br>IBM_TDB | OR               | OFF     | 2007/07/29 08:36 |
| S4    | 0    | S2 and S3  | US-PGPUB;<br>USPAT;<br>USOCR;<br>EPO; JPO;<br>DERWENT;<br>IBM_TDB | OR               | OFF     | 2007/07/29 08:36 |
| S5    | 1426 | ((cluster\$3 or group\$3) with categor\$3 with (structur\$3 or tree or hierarch\$3))                 | US-PGPUB;<br>USPAT;<br>USOCR;<br>EPO; JPO;<br>DERWENT;<br>IBM_TDB | OR               | OFF     | 2007/07/29 08:37 |
| S6    | 10   | S3 and S5  | US-PGPUB;<br>USPAT;<br>USOCR;<br>EPO; JPO;<br>DERWENT;<br>IBM_TDB | OR               | OFF     | 2007/07/29 08:37 |
| S7    | 5    | S6 and @ad<"20010904"  | US-PGPUB;<br>USPAT;<br>USOCR;<br>EPO; JPO;<br>DERWENT;<br>IBM_TDB | OR               | OFF     | 2007/07/29 09:55 |
| S8    | 283  | ((cluster\$3 or group\$3 or class\$9) with categor\$3 with (structur\$3 or tree or hierarch\$3)).ab. | US-PGPUB;<br>USPAT;<br>USOCR;<br>EPO; JPO;<br>DERWENT;<br>IBM_TDB | OR               | OFF     | 2007/07/29 09:54 |

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| S9  | 3308  | ((cluster\$3 or group\$3 or class\$9) with categor\$3 with (structur\$3 or tree or hierarch\$3)) | US-PGPUB; USPAT; USOCR; EPO; JPO; DERWENT; IBM_TDB | OR | OFF | 2007/07/29 09:54 |
| S10 | 26972 | (organizat\$3 with (structur\$3 or tree or hierarch\$3))   | US-PGPUB; USPAT; USOCR; EPO; JPO; DERWENT; IBM_TDB | OR | OFF | 2007/07/29 09:54 |
| S11 | 354   | ( weight\$3 or scor\$3 or rank\$3 ) with vector with categor\$3                                  | US-PGPUB; USPAT; USOCR; EPO; JPO; DERWENT; IBM_TDB | OR | OFF | 2007/07/29 10:41 |
| S12 | 15    | S11 and S9 and @ad<"20010904"  | US-PGPUB; USPAT; USOCR; EPO; JPO; DERWENT; IBM_TDB | OR | OFF | 2007/07/29 11:50 |
| S13 | 6     | S12 and S10  | US-PGPUB; USPAT; USOCR; EPO; JPO; DERWENT; IBM_TDB | OR | OFF | 2007/07/29 09:55 |
| S14 | 15    | ( weight\$3 or scor\$3 or rank\$3 ) and S12  | US-PGPUB; USPAT; USOCR; EPO; JPO; DERWENT; IBM_TDB | OR | OFF | 2007/07/29 11:53 |
| S15 | 15    | S11 and S12  | US-PGPUB; USPAT; USOCR; EPO; JPO; DERWENT; IBM_TDB | OR | OFF | 2007/07/29 11:44 |
| S16 | 1     | S15 and (707/100).ccls.  | US-PGPUB; USPAT; USOCR; EPO; JPO; DERWENT; IBM_TDB | OR | OFF | 2007/07/29 11:49 |

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|-----|------|--|---|----|-----|------------------|
| S17 | 3    | S15 and (707/3).ccls.  | US-PGPUB;<br>USPAT;<br>USOCR;<br>EPO; JPO;<br>DERWENT;<br>IBM_TDB | OR | OFF | 2007/07/29 11:47 |
| S18 | 3    | S15 and (707/5).ccls.  | US-PGPUB;<br>USPAT;<br>USOCR;<br>EPO; JPO;<br>DERWENT;<br>IBM_TDB | OR | OFF | 2007/07/29 11:44 |
| S19 | 675  | (modif\$6 or updat\$3 or chang\$3) with categor\$3 with (structur\$3 or tree or hiearch\$3)                | US-PGPUB;<br>USPAT;<br>USOCR;<br>EPO; JPO;<br>DERWENT;<br>IBM_TDB | OR | OFF | 2007/07/29 11:50 |
| S20 | 158  | S19 and S9   | US-PGPUB;<br>USPAT;<br>USOCR;<br>EPO; JPO;<br>DERWENT;<br>IBM_TDB | OR | OFF | 2007/07/29 11:50 |
| S21 | 1    | S20 and S11 and @ad<"20010904"   | US-PGPUB;<br>USPAT;<br>USOCR;<br>EPO; JPO;<br>DERWENT;<br>IBM_TDB | OR | OFF | 2007/07/29 11:50 |
| S22 | 5590 | (calculat\$3 or comput\$3 or measur\$3) with ( weight\$3 or scor\$3 or rank\$3 ) with vector               | US-PGPUB;<br>USPAT;<br>USOCR;<br>EPO; JPO;<br>DERWENT;<br>IBM_TDB | OR | OFF | 2007/07/29 11:54 |
| S23 | 230  | (calculat\$3 or comput\$3 or measur\$3) with ( weight\$3 or scor\$3 or rank\$3 ) with vector with document | US-PGPUB;<br>USPAT;<br>USOCR;<br>EPO; JPO;<br>DERWENT;<br>IBM_TDB | OR | OFF | 2007/07/29 11:55 |
| S24 | 1    | S23 and S8 and S11   | US-PGPUB;<br>USPAT;<br>USOCR;<br>EPO; JPO;<br>DERWENT;<br>IBM_TDB | OR | OFF | 2007/07/29 11:55 |

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method to automatically generate **category terms** and **structure**. The documents were first transformed to. a. set of **feature vectors**, in which each component ...[ieeexplore.ieee.org/iel5/6927/18671/00861377.pdf](http://ieeexplore.ieee.org/iel5/6927/18671/00861377.pdf) - [Similar pages](#)**[PDF] OCRed document**

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entire re-calculation of the **cluster structure**. Also, in our. genetic work for evolving clusterings, mutation will make. **feature vectors** move in and out of ...[ieeexplore.ieee.org/iel5/9096/28877/01299776.pdf?arnumber=1299776](http://ieeexplore.ieee.org/iel5/9096/28877/01299776.pdf?arnumber=1299776) - [Similar pages](#)[ More results from [ieeexplore.ieee.org](http://ieeexplore.ieee.org) ]**[PDF] LNCS 1910 - Automatic Category Structure Generation and ...**

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in word **cluster map** are then used to select **category terms**. The **structure .... feature vectors** we discarded those words which occur only once in a document. ...[www.springerlink.com/index/0XJGRPCNQL31VAAG.pdf](http://www.springerlink.com/index/0XJGRPCNQL31VAAG.pdf) - [Similar pages](#)**[PDF] A Method of Improving Feature Vector for Web Pages Reflecting the ...**

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link **structure** can be classified into the following two **categories**: ..... the centroid **vector** of a **cluster** on each element of the initial **feature vector**. ...[www.springerlink.com/index/5F8TE0VKKGQWYP91.pdf](http://www.springerlink.com/index/5F8TE0VKKGQWYP91.pdf) - [Similar pages](#)[ More results from [www.springerlink.com](http://www.springerlink.com) ]**[PDF] Mining the Structure of User Activity using Cluster Stability**File Format: PDF/Adobe Acrobat - [View as HTML](#)combination (weighted **vector** sum) of the **feature vectors**, using the user .... This **structure** consists of 15 unique tasks, grouped into 5 task **categories**: ...[www2.parc.com/istl/groups/uir/publications/items/UIR-2002-02-Heer-SIAM-ClusterStability.pdf](http://www2.parc.com/istl/groups/uir/publications/items/UIR-2002-02-Heer-SIAM-ClusterStability.pdf) - [Similar pages](#)**All About ISOC: Conferences - INET**We named the nodes in the **category structure** as "templates. ... We did not select all words of the Web pages for the **feature vector**. ...[www.isoc.org/inet98/proceedings/1x/1x\\_5.htm](http://www.isoc.org/inet98/proceedings/1x/1x_5.htm) - 31k - [Cached](#) - [Similar pages](#)**System for categorizing documents in a linked collection of ...**processing means for applying said classification criteria to **feature vectors** to determine if a document is in a corresponding **category**. ...[www.patentstorm.us/patents/5895470-claims.html](http://www.patentstorm.us/patents/5895470-claims.html) - 26k - [Cached](#) - [Similar pages](#)**[PS] Representing, Learning, and Recognizing Non-Rigid Textures and ...**File Format: Adobe PostScript - [View as Text](#)The **feature vector** associated with a node is an intensity-domain spin image ... is simply a set of **cluster** centers with weights corresponding to relative ...[www-cvr.ai.uiuc.edu/ponce\\_grp/demo/vision/snowbird03\\_abstract.ps](http://www-cvr.ai.uiuc.edu/ponce_grp/demo/vision/snowbird03_abstract.ps) - [Similar pages](#)**[PDF] Su Li f Natural Sciences Article IDú1007i¬1202(2004)03i¬0339i¬04 ...**

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... to reduce the dimensionality of the **category feature vector** space which can ... get the probability **structure** of input **vectors** quickly and its **cluster** ...  
[engine.cqvip.com/content/tn/85480x/2004/009/003/gc54\\_tn6\\_10166072.pdf](http://engine.cqvip.com/content/tn/85480x/2004/009/003/gc54_tn6_10166072.pdf) - [Similar pages](#)

**Method and apparatus for determining and organizing feature ...**

The device of claim 2 wherein the **structure** of said neural network has at least one .....

The **feature vector** is obtained 270 and the appropriate **cluster** is ...

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### 1 [Data clustering: a review](#)

 A. K. Jain, M. N. Murty, P. J. Flynn

September 1999 **ACM Computing Surveys (CSUR)**, Volume 31 Issue 3

**Publisher:** ACM Press

Full text available:  [pdf\(636.24 KB\)](#) Additional Information: [full citation](#), [abstract](#), [references](#), [citations](#), [index terms](#), [review](#)

Clustering is the unsupervised classification of patterns (observations, data items, or feature vectors) into groups (clusters). The clustering problem has been addressed in many contexts and by researchers in many disciplines; this reflects its broad appeal and usefulness as one of the steps in exploratory data analysis. However, clustering is a difficult problem combinatorially, and differences in assumptions and contexts in different communities has made the transfer of useful generic co ...

**Keywords:** cluster analysis, clustering applications, exploratory data analysis, incremental clustering, similarity indices, unsupervised learning

### 2 [Special issue on special feature: Distributional word clusters vs. words for text categorization](#)

Ron Bekkerman, Ran El-Yaniv, Naftali Tishby, Yoad Winter

March 2003 **The Journal of Machine Learning Research**, Volume 3

**Publisher:** MIT Press

Full text available:  [pdf\(176.53 KB\)](#) Additional Information: [full citation](#), [abstract](#), [citations](#), [index terms](#)

We study an approach to text categorization that combines distributional clustering of words and a Support Vector Machine (SVM) classifier. This word-cluster representation is computed using the recently introduced *Information Bottleneck* method, which generates a compact and efficient representation of documents. When combined with the classification power of the SVM, this method yields high performance in text categorization. This novel combination of SVM with word-cluster representation ...

### 3 [Feature-based similarity search in graph structures](#)

 Xifeng Yan, Feida Zhu, Philip S. Yu, Jiawei Han

December 2006 **ACM Transactions on Database Systems (TODS)**, Volume 31 Issue 4

**Publisher:** ACM Press

Full text available:  [pdf\(1.44 MB\)](#) Additional Information: [full citation](#), [abstract](#), [references](#), [index terms](#)

Similarity search of complex structures is an important operation in graph-related applications since exact matching is often too restrictive. In this article, we investigate the issues of *substructure similarity search using indexed features* in graph databases. By transforming the edge relaxation ratio of a query graph into the maximum allowed feature misses, our structural filtering algorithm can filter graphs without performing pairwise similarity computation. It is further shown that ...

**Keywords:** Graph database, complexity, index, similarity search

4 **Special issue on ICML: Coupled clustering: a method for detecting structural correspondence**

Zvika Marx, Ido Dagan, Joachim M. Buhmann, Eli Shamir  
March 2003 **The Journal of Machine Learning Research**, Volume 3

**Publisher:** MIT Press

Full text available:  [pdf\(967.15 KB\)](#) Additional Information: [full citation](#), [abstract](#), [citations](#), [index terms](#)

This paper proposes a new paradigm and a computational framework for revealing equivalencies (analogies) between sub-structures of distinct composite systems that are initially represented by unstructured data sets. For this purpose, we introduce and investigate a variant of traditional data clustering, termed *coupled clustering*, which outputs a configuration of corresponding subsets of two such representative sets. We apply our method to synthetic as well as textual data. Its achievement ...

5 **Research track: Visualizing changes in the structure of data for exploratory feature selection**

Elias Pampalk, Werner Goebel, Gerhard Widmer

August 2003 **Proceedings of the ninth ACM SIGKDD international conference on Knowledge discovery and data mining KDD '03**

**Publisher:** ACM Press

Full text available:  [pdf\(642.44 KB\)](#) Additional Information: [full citation](#), [abstract](#), [references](#), [citations](#), [index terms](#)

Using visualization techniques to explore and understand high-dimensional data is an efficient way to combine human intelligence with the immense brute force computation power available nowadays. Several visualization techniques have been developed to study the cluster structure of data, i.e., the existence of distinctive groups in the data and how these clusters are related to each other. However, only few of these techniques lend themselves to studying how this structure changes if the feature ...

**Keywords:** high-dimensional data, interactive data mining

6 **Content 2: image clustering: Web image clustering by consistent utilization of visual features and surrounding texts**

Bin Gao, Tie-Yan Liu, Tao Qin, Xin Zheng, Qian-Sheng Cheng, Wei-Ying Ma

November 2005 **Proceedings of the 13th annual ACM international conference on Multimedia MULTIMEDIA '05**

**Publisher:** ACM Press

Full text available:  [pdf\(1.23 MB\)](#) Additional Information: [full citation](#), [abstract](#), [references](#), [index terms](#)

Image clustering, an important technology for image processing, has been actively researched for a long period of time. Especially in recent years, with the explosive growth of the Web, image clustering has even been a critical technology to help users digest the large amount of online visual information. However, as far as we know, many previous works on image clustering only used either low-level visual features or surrounding texts, but rarely exploited these two kinds of information in the s ...

**Keywords:** co-clustering, consistency, image processing, spectral graph

**7 Semantic clustering and querying on heterogeneous features for visual data**

 Gholamhosein Sheikholeslami, Wendy Chang, Aidong Zhang  
September 1998 **Proceedings of the sixth ACM international conference on Multimedia MULTIMEDIA '98**

**Publisher:** ACM Press

Full text available:  pdf(1.37 MB) Additional Information: [full citation](#), [references](#), [citations](#), [index terms](#)

**8 Clustering and searching WWW images using link and page layout analysis**

 Xiaofei He, Deng Cai, Ji-Rong Wen, Wei-Ying Ma, Hong-Jiang Zhang  
May 2007 **ACM Transactions on Multimedia Computing, Communications, and Applications (TOMCCAP)**, Volume 3 Issue 2

**Publisher:** ACM Press

Full text available:  pdf(28.98 MB) Additional Information: [full citation](#), [abstract](#), [references](#), [index terms](#)

Due to the rapid growth of the number of digital images on the Web, there is an increasing demand for an effective and efficient method for organizing and retrieving the available images. This article describes iFind, a system for clustering and searching WWW images. By using a vision-based page segmentation algorithm, a Web page is partitioned into blocks, and the textual and link information of an image can be accurately extracted from the block containing that image. The textual informatio ...

**Keywords:** Web mining, image clustering, image search, link analysis

**9 Applications: Fast retrieval of high-dimensional feature vectors in P2P networks using**

 **compact peer data summaries**  
Wolfgang Müller, Andreas Henrich  
November 2003 **Proceedings of the 5th ACM SIGMM international workshop on Multimedia information retrieval MIR '03**

**Publisher:** ACM Press

Full text available:  pdf(378.07 KB) Additional Information: [full citation](#), [abstract](#), [references](#), [citations](#), [index terms](#)

The retrieval facilities of most Peer-to-Peer (P2P) systems are limited to queries based on a unique identifier or a small set of keywords. The techniques used for this purpose are hardly applicable for content-based image retrieval (CBIR) in a P2P network. Furthermore, we will argue that the curse of dimensionality and the high communication overhead prevent the adaptation of multidimensional search trees or fast sequential scan techniques for P2P CBIR. In the present paper we will propose two ...

**10 Description and Analysis: Using web structure for classifying and describing web**

 **pages**  
Eric J. Glover, Kostas Tsioutsiouliklis, Steve Lawrence, David M. Pennock, Gary W. Flake  
May 2002 **Proceedings of the 11th international conference on World Wide Web WWW '02**

**Publisher:** ACM Press

Full text available:  pdf(136.12 KB) Additional Information: [full citation](#), [abstract](#), [references](#), [citations](#), [index terms](#)

The structure of the web is increasingly being used to improve organization, search, and analysis of information on the web. For example, Google uses the text in citing documents

(documents that link to the target document) for search. We analyze the relative utility of document text, and the text in citing documents near the citation, for classification and description. Results show that the text in citing documents, when available, often has greater discriminative and descriptive power than th ...

**Keywords:** SVM, anchortext, classification, cluster naming, entropy based feature extraction, evaluation, web directory, web structure

11 Research track paper: Consistent bipartite graph co-partitioning for star-structured high-order heterogeneous data co-clustering

Bin Gao, Tie-Yan Liu, Xin Zheng, Qian-Sheng Cheng, Wei-Ying Ma  
August 2005 **Proceeding of the eleventh ACM SIGKDD international conference on Knowledge discovery in data mining KDD '05**

**Publisher:** ACM Press

Full text available: [pdf\(560.22 KB\)](#) Additional Information: [full citation](#), [abstract](#), [references](#), [citations](#), [index terms](#)

Heterogeneous data co-clustering has attracted more and more attention in recent years due to its high impact on various applications. While the co-clustering algorithms for two types of heterogeneous data (denoted by pair-wise co-clustering), such as documents and terms, have been well studied in the literature, the work on more types of heterogeneous data (denoted by high-order co-clustering) is still very limited. As an attempt in this direction, in this paper, we worked on a specific case of ...

**Keywords:** co-clustering, consistency, high-order heterogeneous data, spectral graph

12 Spectral clustering for multi-type relational data

Bo Long, Zhongfei (Mark) Zhang, Xiaoyun Wú, Philip S. Yu  
June 2006 **Proceedings of the 23rd international conference on Machine learning ICML '06**

**Publisher:** ACM Press

Full text available: [pdf\(240.84 KB\)](#) Additional Information: [full citation](#), [abstract](#), [references](#), [cited by](#), [index terms](#)

Clustering on multi-type relational data has attracted more and more attention in recent years due to its high impact on various important applications, such as Web mining, e-commerce and bioinformatics. However, the research on general multi-type relational data clustering is still limited and preliminary. The contribution of the paper is three-fold. First, we propose a general model, the collective factorization on related matrices, for multi-type relational data clustering. The model is appli ...

13 Concept features in Re:Agent, an intelligent Email agent

Gary Boone  
May 1998 **Proceedings of the second international conference on Autonomous agents AGENTS '98**

**Publisher:** ACM Press

Full text available: [pdf\(1.07 MB\)](#) Additional Information: [full citation](#), [references](#), [citations](#), [index terms](#)

14 Research sessions: Research 7: Indexing: FIX: feature-based indexing technique for XML documents

Ning Zhang, M. Tamer Özsu, Ihab F. Ilyas, Ashraf Aboulnaga  
September 2006 **Proceedings of the 32nd international conference on Very large data bases VLDB '06**

**Publisher:** VLDB Endowment

Full text available: [pdf\(965.30 KB\)](#) Additional Information: [full citation](#), [abstract](#), [references](#), [index terms](#)

Indexing large XML databases is crucial for efficient evaluation of XML twig queries. In this paper, we propose a feature-based indexing technique, called FIX, based on spectral graph theory. The basic idea is that for each twig pattern in a collection of XML documents, we calculate a vector of features based on its structural properties. These features are used as keys for the patterns and stored in a B<sup>+</sup>tree. Given an XPath query, its feature vector is first calculated and looked up ...

**15 Multimedia and visualization (MV): A pivot-based index structure for combination of feature vectors**

 Benjamin Bustos, Daniel Keim, Tobias Schreck

March 2005 **Proceedings of the 2005 ACM symposium on Applied computing SAC '05**

**Publisher:** ACM Press

Full text available: [pdf\(172.44 KB\)](#) Additional Information: [full citation](#), [abstract](#), [references](#), [index terms](#)

We present a novel indexing schema that provides efficient nearest-neighbor queries in multimedia databases consisting of objects described by multiple feature vectors. The benefits of the simultaneous usage of several (statically or dynamically) weighted feature vectors with respect to retrieval *effectiveness* have been previously demonstrated.

Support for *efficient* multi-feature vector similarity queries is an open problem, as existing indexing methods do not support dynamically p ...

**Keywords:** combination of features, content-based indexing and retrieval, nearest neighbor queries

**16 Semantic annotation and integration: Web taxonomy integration using support vector machines**

 Dell Zhang, Wee Sun Lee

May 2004 **Proceedings of the 13th international conference on World Wide Web WWW '04**

**Publisher:** ACM Press

Full text available: [pdf\(191.33 KB\)](#) Additional Information: [full citation](#), [abstract](#), [references](#), [citations](#), [index terms](#)

We address the problem of integrating objects from a source taxonomy into a master taxonomy. This problem is not only currently pervasive on the web, but also important to the emerging semantic web. A straightforward approach to automating this process would be to train a classifier for each category in the master taxonomy, and then classify objects from the source taxonomy into these categories. In this paper we attempt to use a powerful classification method, Support Vector Machine (SVM), to a ...

**Keywords:** classification, ontology mapping, semantic web, support vector machines, taxonomy integration, transductive learning

**17 Real-time shading**

 Marc Olano, Kurt Akeley, John C. Hart, Wolfgang Heidrich, Michael McCool, Jason L. Mitchell, Randi Rost

August 2004 **ACM SIGGRAPH 2004 Course Notes SIGGRAPH '04**

**Publisher:** ACM Press

Full text available: [pdf\(7.39 MB\)](#) Additional Information: [full citation](#), [abstract](#)

Real-time procedural shading was once seen as a distant dream. When the first version of this course was offered four years ago, real-time shading was possible, but only with one-

of-a-kind hardware or by combining the effects of tens to hundreds of rendering passes. Today, almost every new computer comes with graphics hardware capable of interactively executing shaders of thousands to tens of thousands of instructions. This course has been redesigned to address today's real-time shading capabili ...

**18 Content 2: image clustering: Iteratively clustering web images based on link and attribute reinforcements**

 Xin-Jing Wang, Wei-Ying Ma, Lei Zhang, Xing Li

November 2005 **Proceedings of the 13th annual ACM international conference on Multimedia MULTIMEDIA '05**

**Publisher:** ACM Press

Full text available:  [pdf\(248.02 KB\)](#) Additional Information: [full citation](#), [abstract](#), [references](#), [index terms](#)

Image clustering is an important research topic which contributes to a wide range of applications. Traditional image clustering approaches are based on image content features only, while content features alone can hardly describe the semantics of the images. In the context of Web, images are no longer assumed homogeneous and "flat" distributed but are richly structured. There are two kinds of reinforcements embedded in such data: 1) the reinforcement between attributes of different data types (int ...

**Keywords:** image clustering, iterative reinforcement, link mining

**19 PageCluster: Mining conceptual link hierarchies from Web log files for adaptive Web site navigation**

 Jianhan Zhu, Jun Hong, John G. Hughes

May 2004 **ACM Transactions on Internet Technology (TOIT)**, Volume 4 Issue 2

**Publisher:** ACM Press

Full text available:  [pdf\(280.84 KB\)](#) Additional Information: [full citation](#), [abstract](#), [references](#), [citations](#), [index terms](#)

User traversals on hyperlinks between Web pages can reveal semantic relationships between these pages. We use user traversals on hyperlinks as weights to measure semantic relationships between Web pages. On the basis of these weights, we propose a novel method to put Web pages on a Web site onto different conceptual levels in a link hierarchy. We develop a clustering algorithm called PageCluster, which clusters conceptually-related pages on each conceptual level of the link hierarchy based on th ...

**Keywords:** Link hierarchies, Web site navigation, bibliographic analysis, clustering, conceptual link hierarchies, link similarity

**20 A comparative study for domain ontology guided feature extraction**

Bill B. Wang, R. I. Bob McKay, Hussein A. Abbass, Michael Barlow

February 2003 **Proceedings of the 26th Australasian computer science conference - Volume 16 ACSC '03**

**Publisher:** Australian Computer Society, Inc.

Full text available:  [pdf\(119.73 KB\)](#) Additional Information: [full citation](#), [abstract](#), [references](#), [citations](#), [index terms](#)

We introduced a novel method employing a hierarchical domain ontology structure to extract features representing documents in our previous publication (Wang 2002). All raw words in the training documents are mapped to concepts in a concept hierarchy derived from the domain ontology. Based on these concepts, a concept hierarchy is established for the training document space, using is-a relationships defined in the domain ontology. An optimum concept set may be obtained by searching the concept hi ...

**Keywords:**  $\chi^2$  statistics, KNN algorithm, concept hierarchy, information gain, ontology, principal component analysis, text classification

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